A NOTE ON BURKHOLDER-ROSENTHAL INEQUALITY

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ABSTRACT. Let df be a Hilbert-space-valued martingale difference sequence. The paper is devoted to a new, elementary proof of the estimate

$$\left\| \sum_{k=0}^{\infty} df_k \right\|_p \le C_p \left\{ \left\| \left(\sum_{k=0}^{\infty} \mathbb{E}(|df_k|^2 | \mathcal{F}_{k-1}) \right)^{1/2} \right\|_p + \left\| \left(\sum_{k=0}^{\infty} |df_k|^p \right)^{1/p} \right\|_p \right\},$$
with $C_p = O(p/\ln p)$ as $p \to \infty$.

1. INTRODUCTION

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, filtered by $(\mathcal{F}_n)_{n\geq 0}$, a nondecreasing family of sub- σ -algebras of \mathcal{F} . Assume that f is an adapted martingale, taking values in a certain separable Hilbert space \mathcal{H} with norm $|\cdot|$ and scalar product $\langle \cdot, \cdot \rangle$. Then $df = (df_n)_{n\geq 0}$, the difference sequence of f, is given by $df_0 = f_0$ and $df_n = f_n - f_{n-1}$, $n \geq 1$. We define the conditional square function of f by

$$s(f) = \left[\sum_{k=0}^{\infty} \mathbb{E}(|df_k|^2 | \mathcal{F}_{k-1})\right]^{1/2}$$

(here and below, $\mathcal{F}_{-1} = \mathcal{F}_0$) and use the notation

$$s_n(f) = \left[\sum_{k=0}^n \mathbb{E}(|df_k|^2 | \mathcal{F}_{k-1})\right]^{1/2}, \qquad n = 0, 1, 2, \dots,$$

for the truncated conditional square function of f.

The purpose of this note is to investigate Burkholder-Rosenthal inequality

(1.1)
$$||f||_{p} \leq c_{p} \left(||s(f)||_{p} + \left\| \left(\sum_{k=0}^{\infty} |df_{k}|^{p} \right)^{1/p} \right\|_{p} \right)$$

where $p \geq 2$ and c_p is a constant depending only on p. The special case in which the martingale f is a sum of independent mean-zero random variables forms an important extension of Khintchine inequality and was studied by Rosenthal in the 60's. The proof from [11] gives the constant c_p which grows exponentially in p as $p \to \infty$. Johnson, Schechtman and Zinn [4] refined the reasoning and showed that the optimal order of c_p as $p \to \infty$ (still in the independent case) is $p/\ln p$. Applying difficult isoperimetric techniques, Talagrand [12] extended this statement to the case of independent Banach-space-valued random variables. Using hypercontractivity

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methods, Kwapień and Szulga [7] gave a completely elementary proof of Talagrand's result.

The inequality (1.1) for general real martingales (and some c_p) was established by Burkholder in [1]. The validity of this estimate with $c_p = O(p/\ln p)$ was proved by Hitczenko [5] (see also [6]). This result was further generalized to vector-valued setting by Pinelis [10]. Consult also Nagaev [8] for a yet another approach.

The purpose of this paper is to present a new and elementary proof of (1.1) with $c_p = O(p/\ln p)$. Precisely, we will establish the following statement.

Theorem 1.1. If f is a Hilbert-space-valued martingale, then for $p \ge 4$ we have

(1.2)
$$||f||_{p} \leq C_{p} \left(||s(f)||_{p}^{p} + \left\| \left(\sum_{k=0}^{\infty} |df_{k}|^{p} \right)^{1/p} \right\|_{p}^{p} \right)^{1/p}$$

where

$$C_p = 2\sqrt{2} \left(\frac{p}{4} + 1\right)^{1/p} \left(1 + \frac{p}{\ln(p/2)}\right).$$

In fact, using Davis' decomposition, we will be able to prove a slightly stronger estimate: see (2.10) and Remark 2.5 below.

A few words about the proof are in order. Hitczenko [5], [6] and Pinelis [10] apply the extrapolation method (good λ -inequality) of Burkholder and Gundy, combined with appropriate version of Prokhorov "arcsinh" estimate for martingales. Nagaev [8] first establishes a certain exponential bound for the tail of f and deduces Burkholder-Rosenthal estimate using a standard integration argument. Our approach is entirely different and exploits the properties of a certain special function; this type of proof can be regarded as an application of Burkholder's method (see [2] and [9] for more on the subject).

2. Proof of Theorem 1.1

The starting point is the following technical estimate proved by Kwapień and Szulga [7].

Lemma 2.1. Let $p \ge 4$ and put

$$\eta = \eta(p) := \frac{\ln(p/2)/p}{1 + \ln(p/2)/p}.$$

Then for any $t \geq 0$ we have

(2.1)
$$(1+t\eta)^p - pt\eta \le 1 + \left(\frac{p}{2} - 1\right)t^2 + t^p.$$

We shall require the following vector-valued version of this bound. From now on, we assume that $p \ge 4$ and that $\sigma = \sigma(p) = \eta(p)/\sqrt{2}$.

Lemma 2.2. For any $y, d \in \mathcal{H}$ we have

(2.2)
$$|y + \sqrt{2}\sigma d|^p - p|y|^{p-2} \langle y, \sqrt{2}\sigma d \rangle \le |y|^p + \frac{p}{2} |y|^{p-2} |d|^2 + |d|^p.$$

Proof. The left-hand side can be rewritten in the form $F(\langle y, \sqrt{2}\sigma d \rangle)$, where

$$F(s) = \left| |y|^2 + 2\sigma^2 |d|^2 + 2s \right|^{p/2} - p|y|^{p-2}s, \qquad s \in \mathbb{R}.$$

Now keep |y| and |d| fixed; since the function F is convex, it suffices to prove the estimate for $\langle y, \sqrt{2}\sigma d \rangle = \pm \sqrt{2}\sigma |y| |d|$, i.e. in the case when y and d are linearly

dependent. If $\langle y, \sqrt{2}\sigma d \rangle = \sqrt{2}\sigma |y||d|$, then (2.2) follows directly from (2.1); on the other hand, if $\langle y, \sqrt{2}\sigma d \rangle = -\sqrt{2}\sigma |y||d|$, we have

$$|y + \sqrt{2}\sigma d|^{p} - p|y|^{p-2} \langle y, \sqrt{2}\sigma d \rangle = ||y| - \sqrt{2}\sigma |d||^{p} + p\sqrt{2}\sigma |y|^{p-1}|d|$$

$$\leq ||y| + \sqrt{2}\sigma |d||^{p} - p\sqrt{2}\sigma |y|^{p-1}|d|,$$

so the claim again follows from (2.1).

The key ingredient of the proof is the special function $U: [0, \infty) \times \mathcal{H} \times [0, \infty) \to \mathbb{R}$, given by

$$U(x, y, z) = \begin{cases} (|y|^2 - x^2)^{p/2} - cx^p - z & \text{if } |y| \ge \sqrt{2}x, \\ |y|^p - (2^{p/2} - 1 + c)x^p - z & \text{if } |y| < \sqrt{2}x, \end{cases}$$

where

$$c = p \, 2^{p/2-2} + 1.$$

Let us list some properties of this function.

Lemma 2.3. (i) For any $(x, y, z) \in [0, \infty) \times \mathcal{H} \times [0, \infty)$ we have

(2.3)
$$U(x,y,z) = \min\left\{ \left| |y|^2 - x^2 \right|^{p/2} - cx^p - z, |y|^p - (2^{p/2} - 1 + c)x^p - z \right\}.$$

(ii) For any $x \ge 0$ and $y \in \mathcal{H}$ we have

(2.4)
$$U(x, \sigma y, |y|^p) \le 0.$$

(iii) For all $(x, y, z) \in [0, \infty) \times \mathcal{H} \times [0, \infty)$ we have

(2.5)
$$U(x, y, z) \ge 2^{-p/2} \left[|y|^p - \sigma^p C_p^p (x^p + z) \right].$$

Proof. (i) For fixed $x, z \ge 0$, the function

$$F(s) = s^{p} - (2^{p/2} - 1 + c)x^{p} - z - (|s^{2} - x^{2}|^{p/2} - cx^{p} - z), \qquad s \ge 0,$$

vanishes at $s = \sqrt{2}x$ and is strictly increasing:

$$F'(s) = ps \left[s^{p-2} - |s^2 - x^2|^{(p-2)/2} \operatorname{sgn}(s^2 - x^2) \right].$$

This yields (2.3).

- (ii) This is obvious, since $\sigma \leq 1$.
- (iii) Using the definitions of C_p and $\sigma,$ we see that we must prove the bound

$$U(x, y, z) \ge 2^{-p/2} \left[|y|^p - \left(\frac{p}{4} + 1\right) 2^p (x^p + z) \right]$$

Now, for $|y| < \sqrt{2}x$, we have

$$U(x, y, z) = |y|^p - \left(\frac{p}{4} + 1\right) 2^{p/2} x^p - z \ge 2^{-p/2} \left[|y|^p - \left(\frac{p}{4} + 1\right) 2^p (x^p + z) \right].$$

On the other hand, if $|y| \ge \sqrt{2}x$, then $|y|^2 - x^2 \ge |y|^2/2$ and hence

$$U(x, y, z) \ge 2^{-p/2} \left[|y|^p - 2^{p/2} c x^p - 2^{p/2} z \right],$$

so the majorization is clear.

We turn to the key property of the function U.

Lemma 2.4. For any $x, z \ge 0$, $y \in \mathcal{H}$ and any \mathcal{H} -valued, mean-zero random variable d with $||d||_p < \infty$ we have

(2.6)
$$\mathbb{E}U\left(\sqrt{x^2 + \mathbb{E}|d|^2}, y + \sigma d, z + |d|^p\right) \le U(x, y, z).$$

Proof. We consider three cases separately.

1° The case
$$|y|^2 \leq 2x^2$$
. By (2.3), we have
 $\mathbb{E}U\left(\sqrt{x^2 + \mathbb{E}|d|^2}, y + \sigma d, z + |d|^p\right)$
 $\leq \mathbb{E}|y + \sigma d|^p - (2^{p/2} - 1 + c)(x^2 + \mathbb{E}|d|^2)^{p/2} - z - \mathbb{E}|d|^p$
 $= \mathbb{E}\left\{|y + \sigma d|^p - p|y|^{p-2}\langle y, \sigma d \rangle - |d|^p\right\} - (2^{p/2} - 1 + c)(x^2 + \mathbb{E}|d|^2)^{p/2} - z.$

By (2.2), the expression in the parentheses does not exceed $|y|^p + p|y|^{p-2}|d|^2/2$; furthermore, we have

$$(2^{p/2} - 1 + c)(x^2 + \mathbb{E}|d|^2)^{p/2} \ge (2^{p/2} - 1 + c)\left(x^p + \frac{p}{2}x^{p-2}\mathbb{E}|d|^2\right)$$
$$\ge (2^{p/2} - 1 + c)x^p + \frac{p}{2}2^{(p-2)/2}x^{p-2}\mathbb{E}|d|^2$$
$$\ge (2^{p/2} - 1 + c)x^p + \frac{p}{2}|y|^{p-2}\mathbb{E}|d|^2.$$

Combining these estimates, we obtain

$$\mathbb{E}U\left(\sqrt{x^2 + \mathbb{E}|d|^2}, y + \sigma d, z + |d|^p\right) \le |y|^p - (2^{p/2} - 1 + c)x^p - z,$$

which is precisely the desired bound.

2° The case $2x^2 < |y|^2 \le 2(x^2 + \mathbb{E}|d|^2)$. We start as previously: by (2.3) and then (2.2),

$$\mathbb{E}U\left(\sqrt{x^2 + \mathbb{E}|d|^2}, y + \sigma d, z + |d|^p\right) \\ \leq |y|^p + \frac{p}{2}|y|^{p-2}\mathbb{E}|d|^2 - (2^{p/2} - 1 + c)(x^2 + \mathbb{E}|d|^2)^{p/2} - z.$$

The latter expression decreases as $\mathbb{E}|d|^2$ increases; indeed, the function

$$F(s) = |y|^{p} + \frac{p}{2}|y|^{p-2}s - (2^{p/2} - 1 + c)(x^{2} + s)^{p/2} - z, \qquad s \ge \frac{|y|^{2}}{2} - x^{2},$$
 fies

satisfies

$$F'(s) \le \frac{p}{2} \left[|y|^{p-2} - 2^{p/2-1} (x^2 + s)^{p/2-1} \right] \le 0.$$

In consequence, we have

$$\begin{split} \mathbb{E}U\left(\sqrt{x^2 + \mathbb{E}|d|^2}, y + \sigma d, z + |d|^p\right) \\ &\leq F\left(\frac{|y|^2}{2} - x^2\right) \\ &= \frac{p}{2}|y|^{p-2}\left(\frac{|y|^2}{2} - x^2\right) - (c-1)\left(\frac{y^2}{2}\right)^{p/2} - z \\ &= -\frac{p}{2}|y|^{p-2}x^2 - z \\ &= \left(\frac{|y|^2}{2}\right)^{p/2-1}x^2 - \left(\frac{p}{2} + 2^{1-p/2}\right)|y|^{p-2}x^2 - z \\ &\leq \left(\frac{|y|^2}{2}\right)^{p/2} - c\left(\frac{|y|^2}{2}\right)^{p/2-1}x^2 - z \\ &\leq (|y|^2 - x^2)^{p/2} - cx^p - z, \end{split}$$

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and we are done.

3° The case $|y|^2 > 2(x^2 + \mathbb{E}|d|^2)$. Here the reasoning is a bit more complicated. First we show the pointwise estimate

(2.7)
$$\begin{aligned} & \left| |y + \sigma d|^2 - x^2 - \mathbb{E} |d|^2 \right|^{p/2} - p \left| |y|^2 - x^2 - \mathbb{E} |d|^2 \right|^{p/2-1} \langle y, \sigma d \rangle \\ & \leq \left(\left| |y|^2 - x^2 - \mathbb{E} |d|^2 \right|^{1/2} + \sqrt{2} \sigma |d| \right)^p - p \left| |y|^2 - x^2 - \mathbb{E} |d|^2 \right|^{(p-1)/2} \sqrt{2} \sigma |d|. \end{aligned}$$

In fact, we will establish a slightly stronger inequality:

$$\begin{aligned} ||y + \sigma d|^2 - x^2 - \mathbb{E}|d|^2 |^{p/2} - p ||y|^2 - x^2 - \mathbb{E}|d|^2 |^{p/2-1} \langle y, \sigma d \rangle \\ &\leq \left(|y|^2 - x^2 - \mathbb{E}|d|^2 + \sigma^2 |d|^2 + 2\sqrt{2} ||y|^2 - x^2 - \mathbb{E}|d|^2 |^{1/2} \sigma |d| \right)^{p/2} \\ &- p ||y|^2 - x^2 - \mathbb{E}|d|^2 |^{(p-1)/2} \sqrt{2} \sigma |d|. \end{aligned}$$

To do this, divide throughout by $||y|^2 - x^2 - \mathbb{E}|d|^2|^{p/2}$ and substitute

$$A^{2} = \frac{|y|^{2} - x^{2} - \mathbb{E}|d|^{2} + \sigma^{2}|d|^{2}}{|y|^{2} - x^{2} - \mathbb{E}|d|^{2}}, \quad Y = \frac{y}{||y|^{2} - x^{2} - \mathbb{E}|d|^{2}|^{1/2}}$$

and

$$D = \frac{d}{||y|^2 - x^2 - \mathbb{E}|d|^2|^{1/2}}.$$

The estimate becomes

(2.8)
$$|A^2 + 2\langle Y, \sigma D\rangle|^{p/2} - p\langle Y, \sigma D\rangle \le |A^2 + 2\sqrt{2}\sigma|D||^p - p\sqrt{2}\sigma|D|.$$

However, the reasoning presented in the proof of (2.2) gives

$$\left|A^{2}+2\langle Y,\sigma D\rangle\right|^{p/2}-p\langle Y,\sigma D\rangle\leq\left(A^{2}+2\sigma|Y||D|\right)^{p/2}-p\sigma|Y||D|.$$

It suffices to use the bounds $|Y| \leq \sqrt{2}$ and $A^2 \geq 1$ to obtain (2.8), because the function $s \mapsto (A^2 + 2s)^{p/2} - ps$ is increasing on $[0, \infty)$. Thus (2.7) follows. We turn to (2.6): applying (2.3), we get

$$\begin{split} &\mathbb{E}U\left(\sqrt{x^{2}+\mathbb{E}|d|^{2}},y+\sigma d,z+|d|^{p}\right)\\ &\leq \mathbb{E}\left||y+\sigma d|^{2}-x^{2}-\mathbb{E}|d|^{2}\right|^{p/2}-c(x^{2}+\mathbb{E}|d|^{2})^{p/2}-z-\mathbb{E}|d|^{p}\\ &= \mathbb{E}\left\{\left||y+\sigma d|^{2}-x^{2}-\mathbb{E}|d|^{2}\right|^{p/2}-p\left||y|^{2}-x^{2}-\mathbb{E}|d|^{2}\right|^{p/2-1}\langle y,\sigma d\rangle\right\}\\ &-c(x^{2}+\mathbb{E}|d|^{2})^{p/2}-z-\mathbb{E}|d|^{p}\\ &\leq \mathbb{E}\left\{\left(\left||y|^{2}-x^{2}-\mathbb{E}|d|^{2}\right|^{1/2}+\sqrt{2}\sigma|d|\right)^{p}-p\left||y|^{2}-x^{2}-\mathbb{E}|d|^{2}\right|^{(p-1)/2}\sqrt{2}\sigma|d|\right\}\\ &-cx^{p}-z-\mathbb{E}|d|^{p}. \end{split}$$

Now we apply (2.2) (in the real case) to obtain

$$\mathbb{E}U\left(\sqrt{x^{2} + \mathbb{E}|d|^{2}}, y + \sigma d, z + |d|^{p}\right)$$

$$\leq \left||y|^{2} - x^{2} - \mathbb{E}|d|^{2}\right|^{p/2} + \frac{p}{2}\left||y|^{2} - x^{2} - \mathbb{E}|d|^{2}\right|^{p/2 - 1} \mathbb{E}|d|^{2} - cx^{p} - z$$

$$\leq \left||y|^{2} - x^{2}\right|^{p/2} - cx^{p} - z = U(x, y, z).$$

This completes the proof.

Proof of (1.2). It suffices to prove that for any nonnegative integer n,

(2.9)
$$\mathbb{E}|f_n|^p \le C_p^p \mathbb{E}\left(s_n^p(f) + \sum_{k=0}^n |df_k|^p\right).$$

Of course, we may assume that df_0, df_1, \ldots, df_n (and hence also f_n) belong to L^p , since otherwise there is nothing to prove. The key observation is that the process

$$\left(U\left(s_n(f),\sigma f_n,\sum_{k=0}^n |df_k|^p\right)\right)_{n\geq 0}$$

is a supermartingale with respect to $(\mathcal{F}_n)_{n\geq 0}$. Indeed, the integrability follows from the above assumption on df; furthermore, for any $n\geq 0$ we have

$$\mathbb{E}\left[U\left(s_{n+1}(f),\sigma f_{n+1},\sum_{k=0}^{n+1}|df_k|^p\right)\middle|\mathcal{F}_n\right]$$
$$=\mathbb{E}\left[U\left(\sqrt{s_n^2(f)+\mathbb{E}(|df_{n+1}|^2|\mathcal{F}_n)},\sigma f_n+\sigma df_{n+1},\sum_{k=0}^{n}|df_k|^p+|df_{n+1}|^p\right)\middle|\mathcal{F}_n\right],$$

which does not exceed $U(s_n(f), \sigma f_n, \sum_{k=0}^n |df_k|^p)$, by (2.6) applied conditionally with respect to \mathcal{F}_{n-1} . Next, we have $U(s_0(f), \sigma f_0, |df_0|^p) \leq 0$, in view of (2.4). Combining these two facts with (2.5) yields the claim:

$$\mathbb{E}\left[|f_n|^p - C_p^p\left(s_n^p(f) + \sum_{k=0}^n |df_k|^p\right)\right] \le \frac{2^{p/2}}{\sigma^p} \mathbb{E}U\left(s_n(f), \sigma f_n, \sum_{k=0}^n |df_k|^p\right) \le 0. \quad \Box$$

Remark 2.5. Using Davis' decomposition (see e.g. Davis [3] or Burkholder [1]), one can deduce a slightly stronger form of (1.2). Namely, for all f as in the statement of Theorem 1.1 and $p \ge 4$ we have

(2.10)
$$||f||_p \le 2C_p \left(||s(f)||_p^p + ||df^*||_p^p \right)^{1/p}$$

where $df^* = \sup_{n\geq 0} |df_n|$. Indeed, fix a martingale f and consider the random variables $d'_n = df_n 1_{\{|df_n| \leq 2df^*_{n-1}\}}, d''_n = df_n 1_{\{|df_n| \geq 2df^*_{n-1}\}}, n = 0, 1, 2, \ldots$ Here, as usual, $df^*_{-1} \equiv 0$ and $df^*_n = \max_{0\leq k\leq n} |df_k|$. Note that on the set $\{|df_n| \geq 2df^*_{n-1}\}$ we have

$$(2^p - 1)|d_n''|^p + (2df_{n-1}^*)^p \le (2|df_n|)^p \le (2df_n^*)^p$$

which implies

$$\sum_{k=0}^{n} |d_k''|^p \le \frac{2^p}{2^p - 1} (df_n^*)^p, \qquad n = 0, \, 1, \, 2, \, \dots$$

Next, observe that for any n,

$$\mathbb{E}\sum_{k=0}^{n} |d'_{k}|^{p} \leq \mathbb{E}\sum_{k=0}^{n} |df_{k}|^{2} (2df_{k-1}^{*})^{p-2}$$

$$= \mathbb{E}\sum_{k=0}^{n} \mathbb{E}(|df_{k}|^{2}|\mathcal{F}_{k-1}) (2df_{k-1}^{*})^{p-2}$$

$$\leq \mathbb{E}s_{n}^{2}(f) (2df_{n}^{*})^{p-2}$$

$$\leq \frac{2}{p} ||s_{n}(f)||_{p}^{p} + \frac{p-2}{p} ||2df_{n}^{*}||_{p}^{p},$$

where in the last line we have exploited Young's inequality. Combining the above estimates for the sums of d'_n and d''_n we get

$$\mathbb{E}\sum_{k=0}^{n} |df_{k}|^{p} \leq \frac{2}{p} ||s_{n}(f)||_{p}^{p} + 2^{p} \left(\frac{1}{2^{p}-1} + \frac{p-2}{p}\right) ||df_{n}^{*}||_{p}^{p}$$
$$\leq \frac{2}{p} ||s_{n}(f)||_{p}^{p} + 2^{p} ||df_{n}^{*}||_{p}^{p}.$$

Plugging this into (2.9) and using the fact that n is an arbitrary nonnegative integer, we obtain (2.10).

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